









#### 4. Conclusions

In this paper, a new Thresholding Eigenvalues based to identify approach is proposed the subspace dimension of the hyperspectral image. The simulation results of the timing analysis and subspace analysis of proposed approach was computed form both synthetic and real dataset. TE approach computes subspace dimensions accurately and comparatively in less time. TE approach can be best suitable for real-time applications.

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